# Machine Learning for Author disambiguation

Gilles Louppe

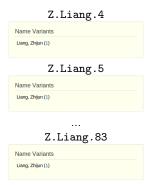
CERN

March 2, 2015

### Motivation

For each author, group together all his publications, and only those.

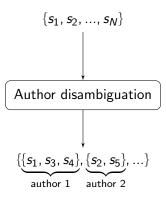






No more No less

But all and only the correct ones



# Spread of the problem

As extracted from claimed publications in INSPIRE,

- Authors have on average 2.06 name variants (synonyms)
   Eg.: Doe, John; Doe, J.
- Unique name variants are shared on average by 1.04 authors (homonyms)

Clustering on exact full names or last name + first initial, should yield very good results on average.

But, disambiguation issues are expected to amplify with the rise of Asian researchers: Caucasian names (now representative of INSPIRE authors) are almost never ambiguous, while Asian names are very often.

### A Preon Model With Family Replication From a D=6, N=2 Supergravity Theory

Hitoshi Nishino, Jogesh C. Pati, S.James Gates, Jr. (Maryland U.)

Dec 1984 - 15 pages

Phys.Lett. B154 (1985) 363 DOI: 10.1016/0370-2693(85)90410-1 MDDP-PP-85-125

### Two Loop Finite Temperature Effective Potential Wess-zumino Model

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Mar 1985 - 22 pages

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Dec 23, 2013 - 6 pages

#### Phys.Rev.Lett. 112 (2014) 131302 (2014-04-02)

DOI: 10.1103/PhysRevLett.112.131302 e-Print: arXiv:1312.6645 [astro-ph.CO] | PDE Experiment: POLARBEAR

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2008 - 1 pages

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### Supergravity in d=9 and Its Coupling to Noncompact $\sigma$ Model

S.J. Gates, Jr. (ICTP, Trieste & Maryland U.), H. Nishino, E. Sezgin (ICTP, Trieste)

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### X Different authors

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Feb 1985 - 14 pages

Phys.Lett. B155 (1985) 421 DOI: 10.1016/0370-2693(85)91598-9 MPI-PAE/PTh 14/85

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✓ Same authors

# Learning from data

- Manual disambiguation is long and difficult, even for experienced curators.
- Couldn't we automatically find a set of rules to disambiguate two signatures?

$$\phi(s_1, s_2) = 
\begin{cases}
0 & \text{if } s_1 \text{ and } s_2 \text{ belong to the same author,} \\
1 & \text{otherwise.} 
\end{cases}$$

This is a machine learning task called supervised learning.

# Supervised learning

- The inputs are random variables  $X = X_1, ..., X_p$ ;
- The output is a random variable Y.
- Data comes as a finite learning set

$$\mathcal{L} = \{(\mathbf{x}_i, \mathbf{y}_i) | i = 0, \dots, N-1\},\$$

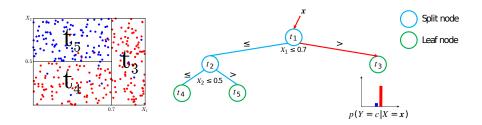
where  $\mathbf{x}_i \in \mathcal{X} = \mathcal{X}_1 \times ... \times \mathcal{X}_p$  and  $\mathbf{y}_i \in \mathcal{Y}$  are randomly drawn from  $P_{\mathbf{X},\mathbf{Y}}$ .

```
E.g., : (\mathbf{x}_i, \mathbf{y}_i) = ((\text{name sim.} = 0.7, \text{title sim.} = 0.3, ...), \text{same authors}) (\mathbf{x}_j, \mathbf{y}_j) = ((\text{name sim.} = 0.1, \text{title sim.} = 0.5, ...), \text{different authors})
```

• The goal is to find a model  $\varphi_{\mathcal{L}}: \mathcal{X} \mapsto \mathcal{Y}$  minimizing

$$Err(\varphi_{\mathcal{L}}) = \mathbb{E}_{X,Y}\{L(Y, \varphi_{\mathcal{L}}(X))\}.$$

# Decision trees [L. Breiman, 1984]

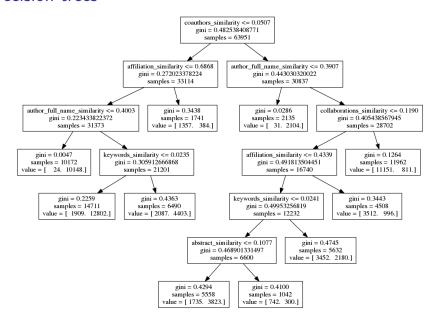


- Heterogeneous data
- Non-parametric model (detect non-linear interactions)
- Easily interpretable
- But prone to overfitting (high variance)



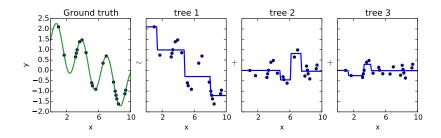
 ${\tt sklearn.tree.DecisionTreeClassifier|Regressor}$ 

### Decision trees

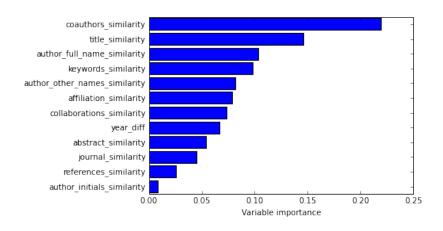


# Gradient Boosted Regression Trees [J. Friedman, 1999]

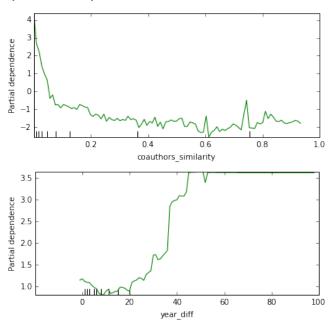
- Ensemble of regression trees approximating the (negative) gradient of a loss function
- Each tree is a successive gradient descent step
- Low bias and low variance

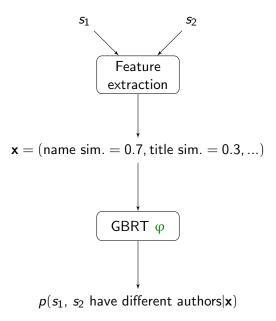


# Variable importances



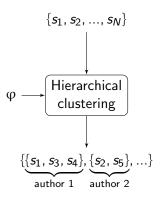
# Partial dependence plots



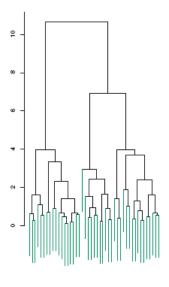


# Disambiguation as a clustering problem

- Author disambiguation = clustering signatures that belong to the same author.
- Using our model  $\phi$ , the probability that two signatures belong to different authors can be used as a (pseudo) distance metric.



# Hierarchical clustering



- General family of clustering algorithms that build nested clusters by merging them successively.
- This hierarchy of clusters is represented as a tree (or dendrogram).
- The root of the tree is the unique cluster that gathers all the samples, the leaves being the clusters with only one sample.

### Issues

- The complexity of hierarchical clustering is  $O(N^2)$ . For  $N=10^7$  signatures, this is impractical. Solution: pre-cluster into blocks all signatures with the same last name + first initial, then cluster each of these blocks.
- How do you set the cut-off threshold?
   Solution: using training data (e.g., claimed signatures), pick the threshold that locally maximizes some criterion.

### **Evaluation**

Protocol: Use the claimed signatures (about 1M) to form ground truth clusters. Keep 10% as a training set to find model parameters, and 90% as a test set for evaluation.

$$B^{3} \text{ Precision} = \mathbb{E}_{s} \{ \frac{|\hat{C}(s) \cap C(s)|}{|\hat{C}(s)|} \}$$
 (1)

$$B^{3} \operatorname{Recall} = \mathbb{E}_{s} \{ \frac{|\hat{C}(s) \cap C(s)|}{|C(s)|} \}$$
 (2)

$$B^3$$
 F-score =  $\frac{2 \times \text{Precision} \times \text{Recall}}{\text{Precision} + \text{Recall}}$  (3)

where C(s) (resp.,  $\hat{C}(s)$ ) is the true (resp., predicted) set of signatures to which s belongs.

# Results

Method	B <sup>3</sup> F-score
Full name	0.8183
Last name $+$ First initial	0.9403
Current prototype	0.9701

# On-going improvements

- Better evaluation metrics.
- Better exploitation of the training data (e.g., for setting the thresholds, for pre-initializing known clusters, etc).
- Evaluate alternative input features, supervised learning algorithms and clustering algorithms.
- Limit model complexity to avoid overfitting and speedup the procedure.
- Deployment.